




# SLAM

# Algorithms

EDAA - G06

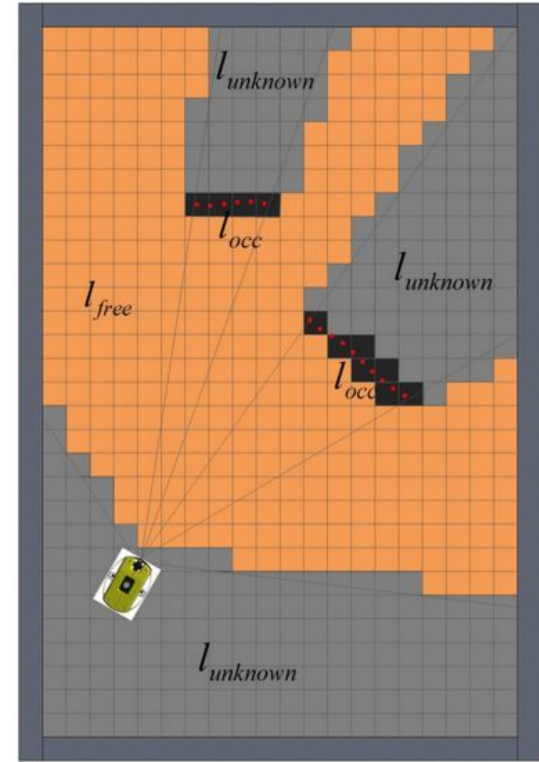
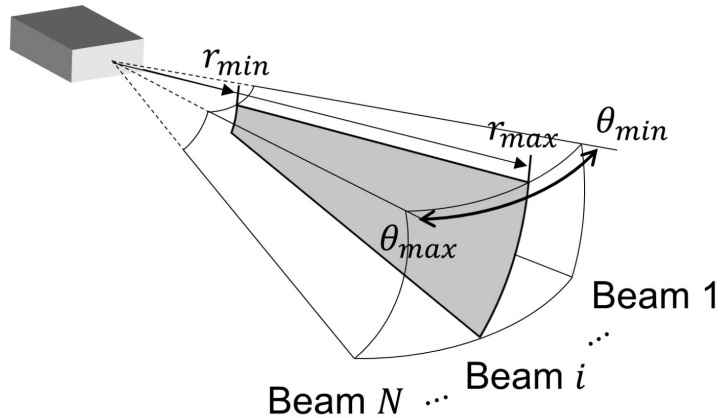
**Tiago Duarte**  
João Costa — João Martins  
Henrique Ribeiro



# 3D Mapping

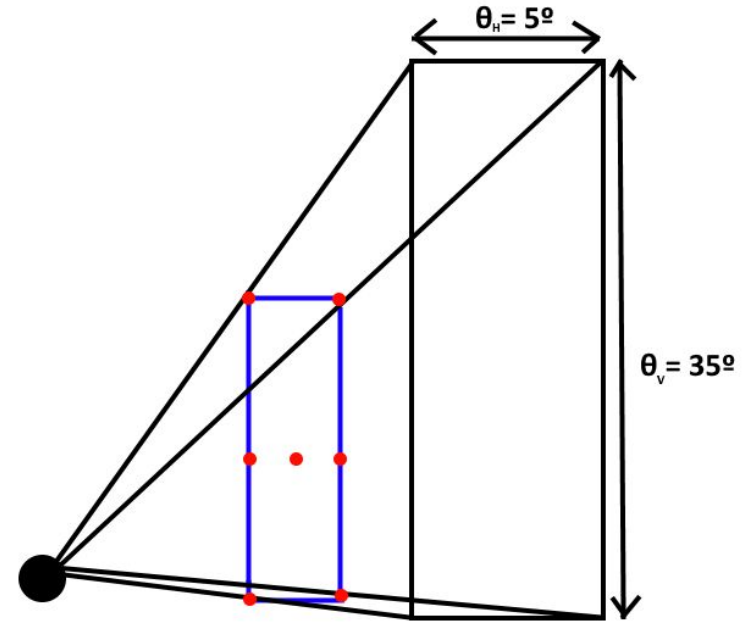
# Sonar Data & 2D Mapping

- Sonar rotates around itself
- Sends/measures waves in a cone
- Each beam has multiple intensities across several intervals

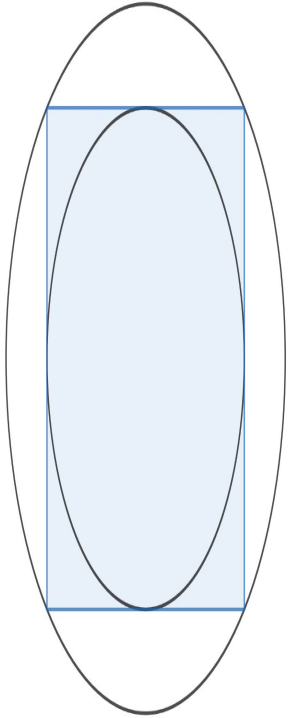


# 3D - How to Cover a Volume?

- Estimate covered cells through 3D raycasts
- Choose destination points to achieve maximum coverage
- Our algorithm, adapted from Fula[0] will cast to 7 points



# Approximating the Sonar as a Rectangle



**Problem:** If we update the bounding rect of the 5° sonar angle, we would map possibly occupied space as empty

**Solution:** By using an angle reduced by a constant factor, it's bounding box will be circumscribed within the origin sonar's beam width

# OctoMap Coverage

The sonar's range is limited to 5m:

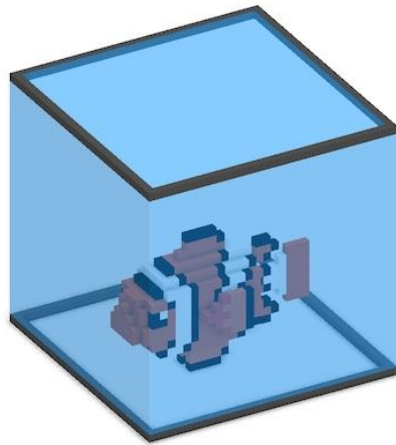
7 raycasts will be sufficient

Let's assume a larger range:

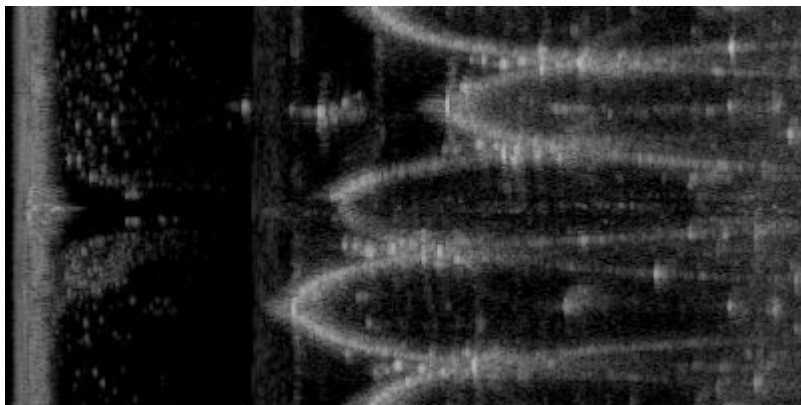
The entire volume **will not be updated**

We can increase the number of raycasts:

**Tradeoff** between **efficiency** and **precision**



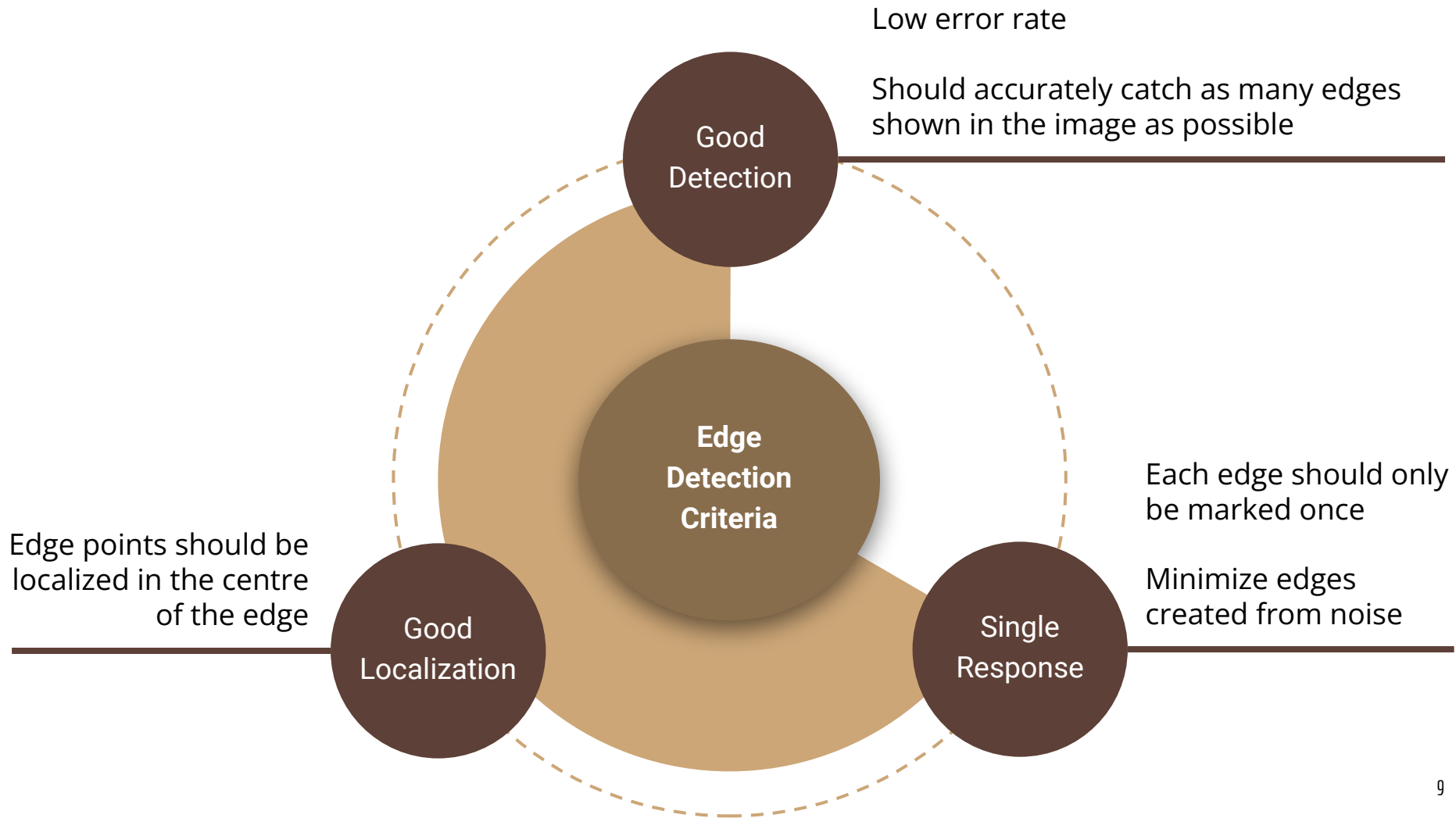
# Range to First Feature



- I. Ignore first few measurements
- II. Apply an intensity threshold
- III. Apply an edge detector
- IV. Select first value higher than a dynamic threshold  $T_d$

# Edge Detection





# Canny Edge Detection

## Image Smoothing

Smooth image using a Gaussian function, with width  $\sigma$

## Differentiation

Calculate the magnitude and angle of the gradient

## Non-Maximum Suppression

Reduce thickness of edges and localize their centre

## Thresholding & Hysteresis

Filter relevant edges using two thresholds and reduce streaking

# Canny results



# [1] Image Smoothing

Convert the image into **grayscale**, then **convolve** the image with a **Gaussian filter** to smooth noise



$$\mathbf{B} = \frac{1}{159} \begin{bmatrix} 2 & 4 & 5 & 4 & 2 \\ 4 & 9 & 12 & 9 & 4 \\ 5 & 12 & 15 & 12 & 5 \\ 4 & 9 & 12 & 9 & 4 \\ 2 & 4 & 5 & 4 & 2 \end{bmatrix}$$

Usually, a 5x5 kernel is used, but its standard deviation can usually be altered

In the kernel above,  $\sigma = 1$

## [2] Differentiation

Calculate the first derivative in the horizontal  $G_x$  and vertical  $G_y$  directions using an **edge detector operator** such as:

- Robert's Cross
- Prewitt's Operator
- Sobel's Operator

$$\begin{bmatrix} +1 & 0 \\ 0 & -1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 0 & +1 \\ -1 & 0 \end{bmatrix}$$

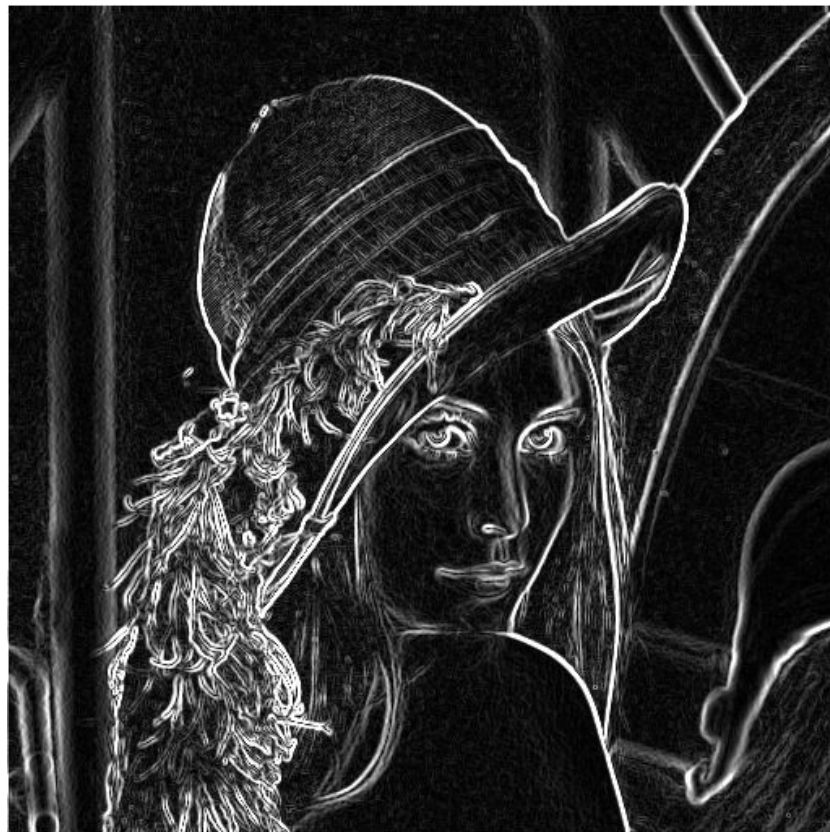
$$\mathbf{G}_x = \begin{bmatrix} +1 & 0 & -1 \\ +2 & 0 & -2 \\ +1 & 0 & -1 \end{bmatrix}$$
$$\mathbf{G}_y = \begin{bmatrix} +1 & +2 & +1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

## [2] Differentiation

To characterize the gradient, two matrixes are created to store its **Intensity** and **Angle**.

$$\mathbf{G} = \sqrt{\mathbf{G}_x^2 + \mathbf{G}_y^2}$$

$$\Theta = \text{atan2}(\mathbf{G}_y, \mathbf{G}_x)$$



# [3] Non-Maximum Suppression

To ensure good localization, we must apply **edge thinning** to remove unwanted points, ideally resulting in one-pixel edges

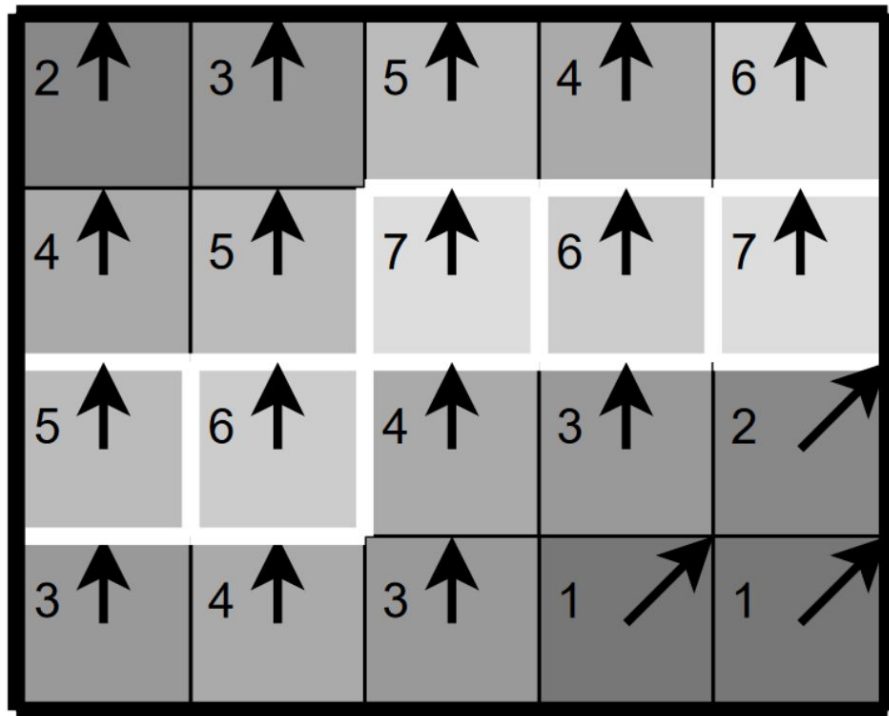
To achieve this, Non-Maximum Suppression removes all points that are not the local maxima of its respective edge

A simple implementation discretizes the gradient's angle into an 8-connectec neighbourhood, and then compare its intensity with the two cells in its direction

If the cell's value is higher, its intensity is kept, otherwise the value is discarded

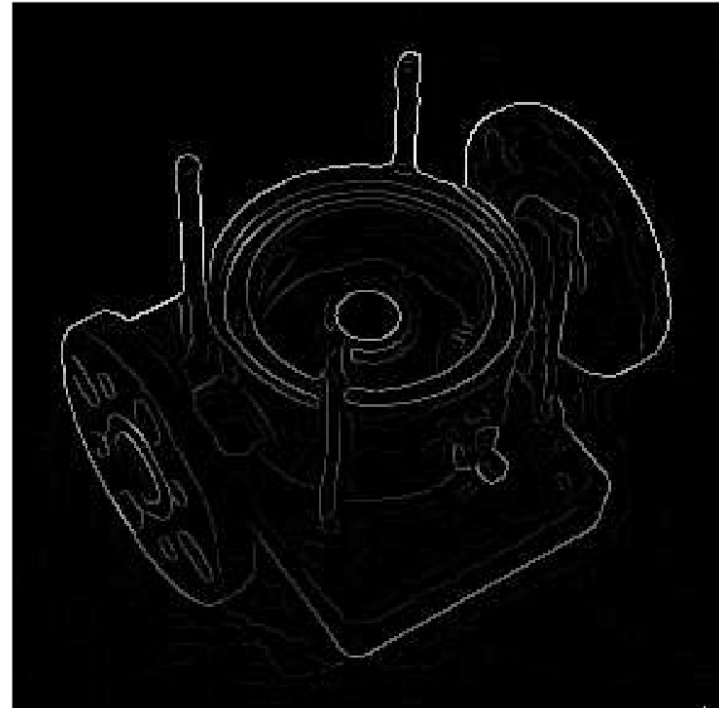


# [3] Non-Maximum Suppression





# [3] Non-Maximum Suppression



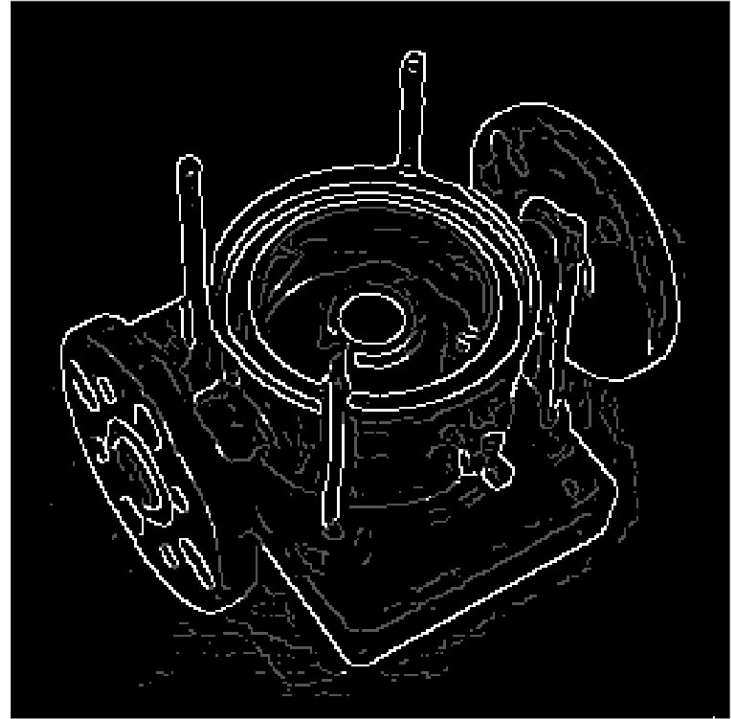
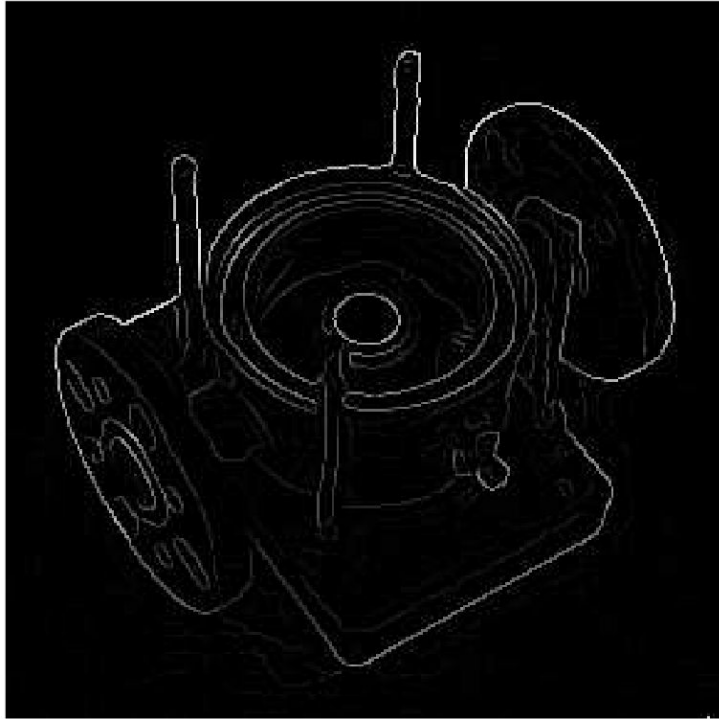
# [4] Threshold & Hysteresis

A single threshold limit usually causes edge values to fluctuate above and below this value, making the line appear broken, called '**streaking**'.

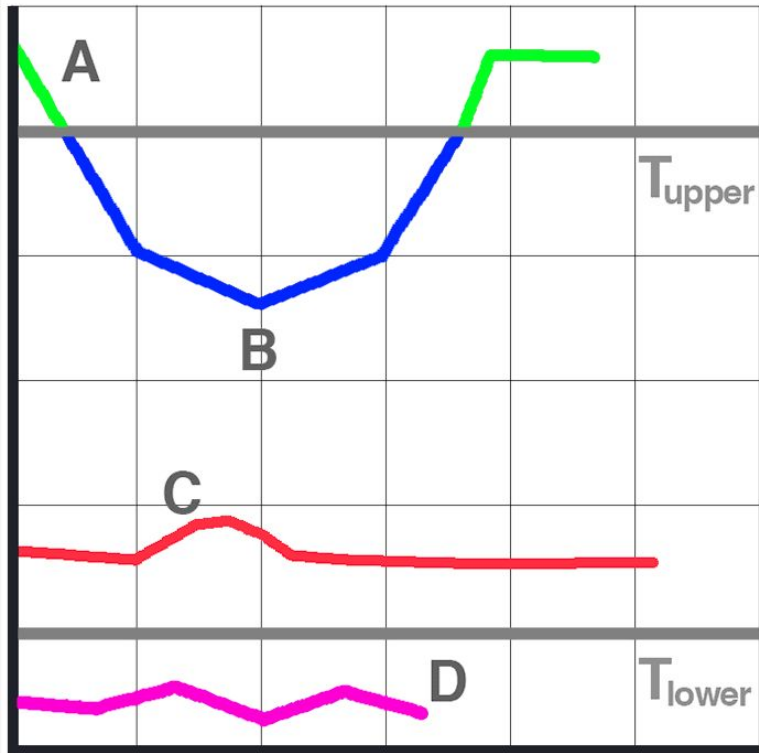
By using two thresholds, we can classify a pixel in 3 categories:

Upper Threshold	<b>Strong</b>	Definitely an edge
	<b>Weak</b>	Edge candidates
Lower Threshold	<b>Ignored</b>	Never an edge

# Strong vs Weak edges



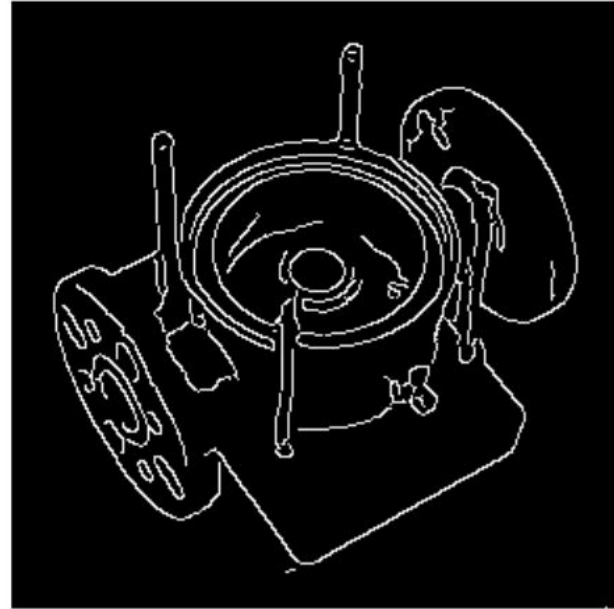
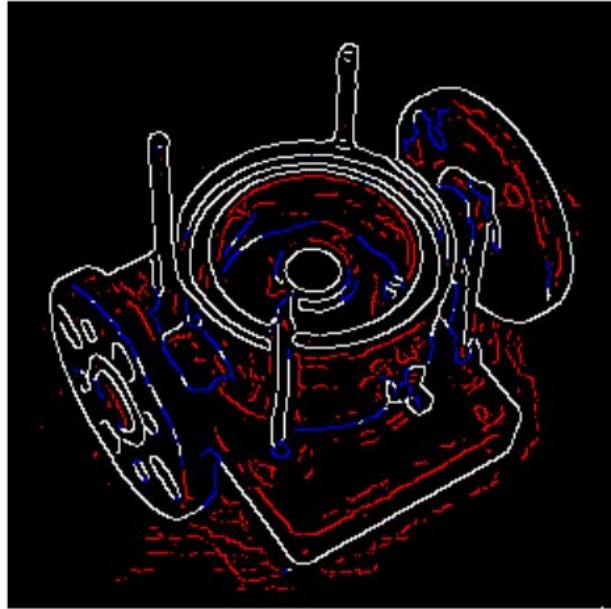
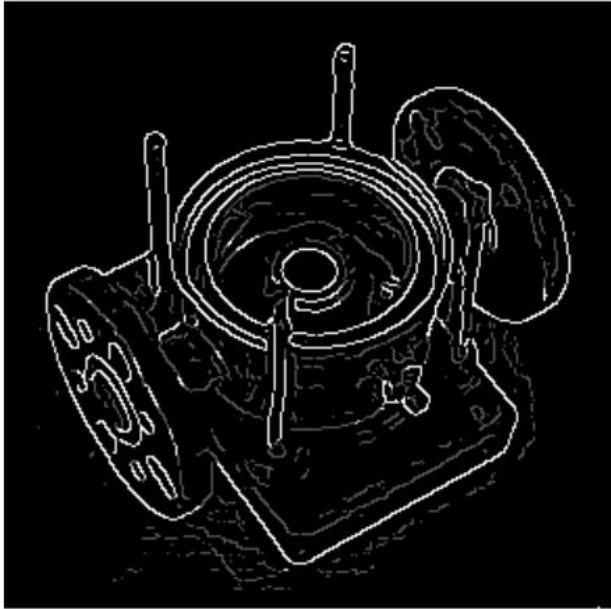
# Threshold & Hysteresis



A	Always an Edge
B	Edge
C	Not an edge
D	Never an edge

Weak edges are only kept if they are connected to a strong edge

# Hysteresis Result



How to determine if a weak pixel belongs to an edge with at least one strong pixel?



# Connected Component Labelling or Blob Analysis

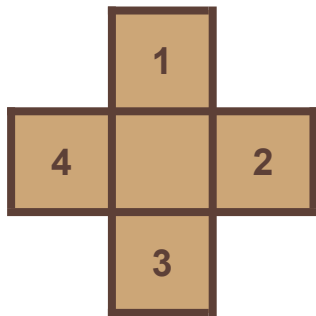


# Connected Component Labelling

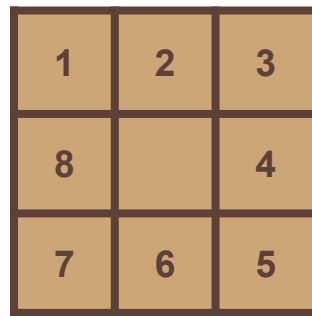
## Goal

- Label each connected region in a binary image, often called blobs, with the same unique label

## Different Connectivities



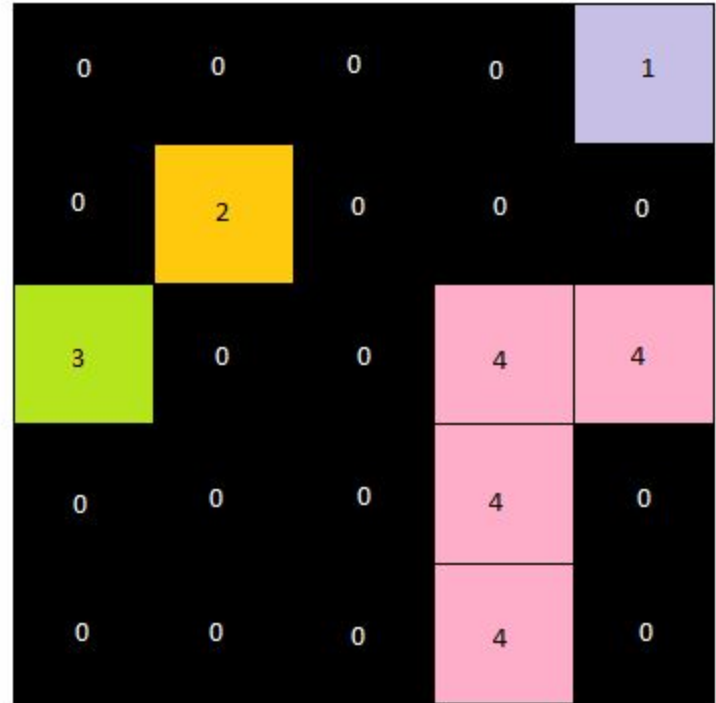
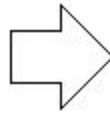
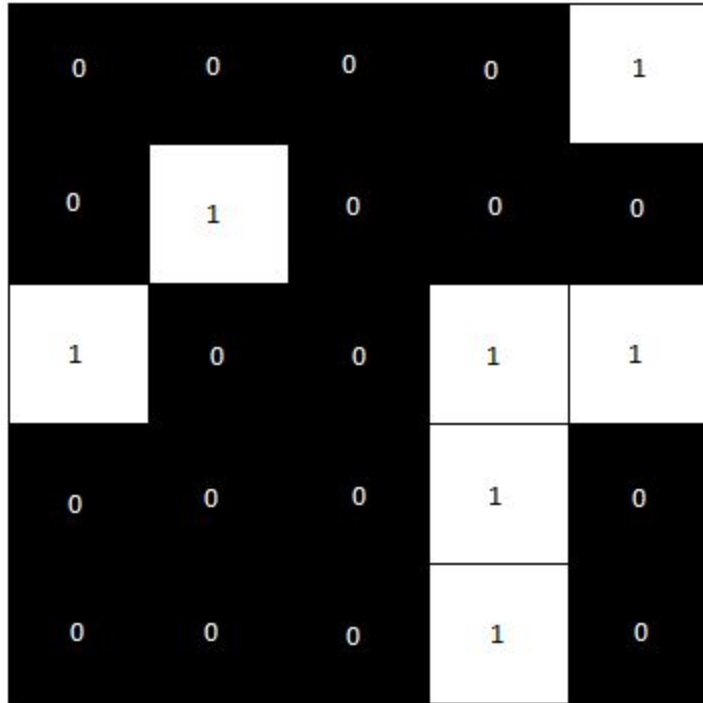
4-way  
connectivity



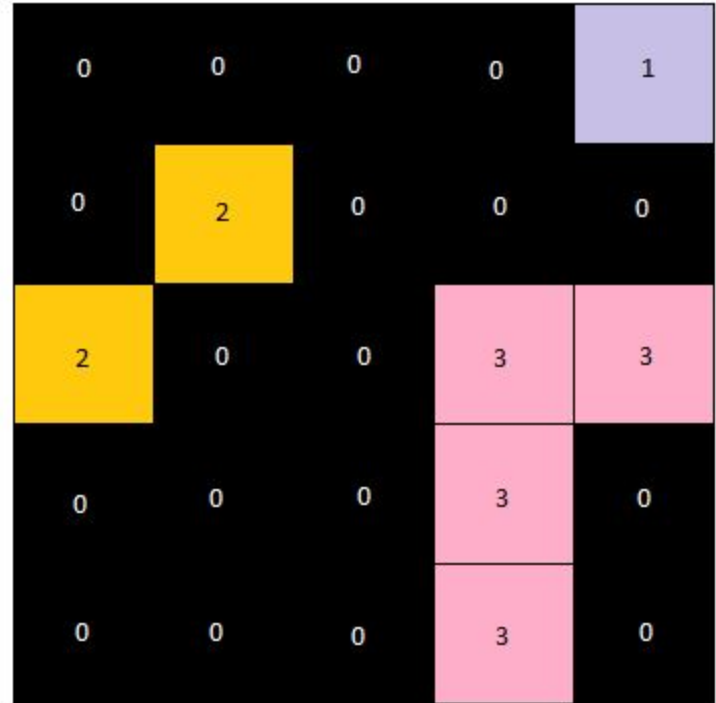
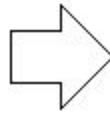
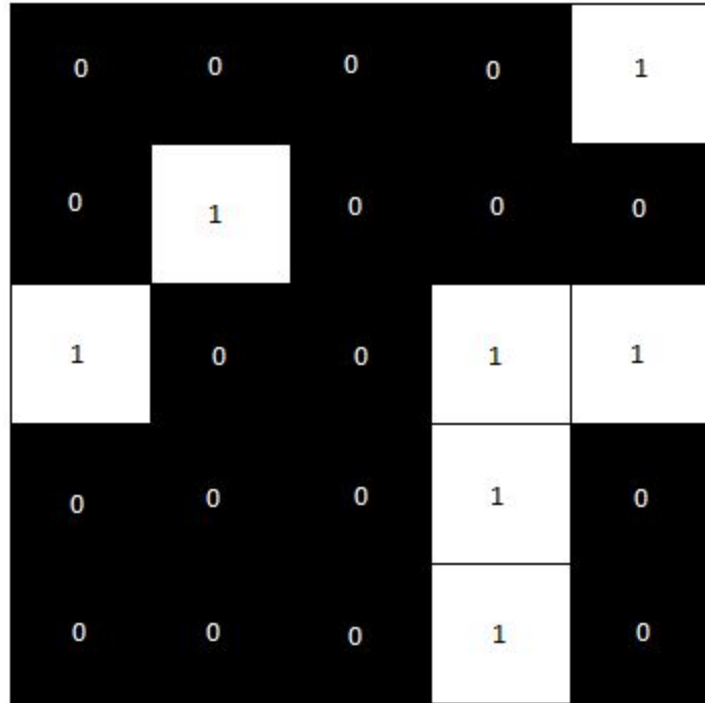
8-way  
connectivity



# 4-Way Connectivity

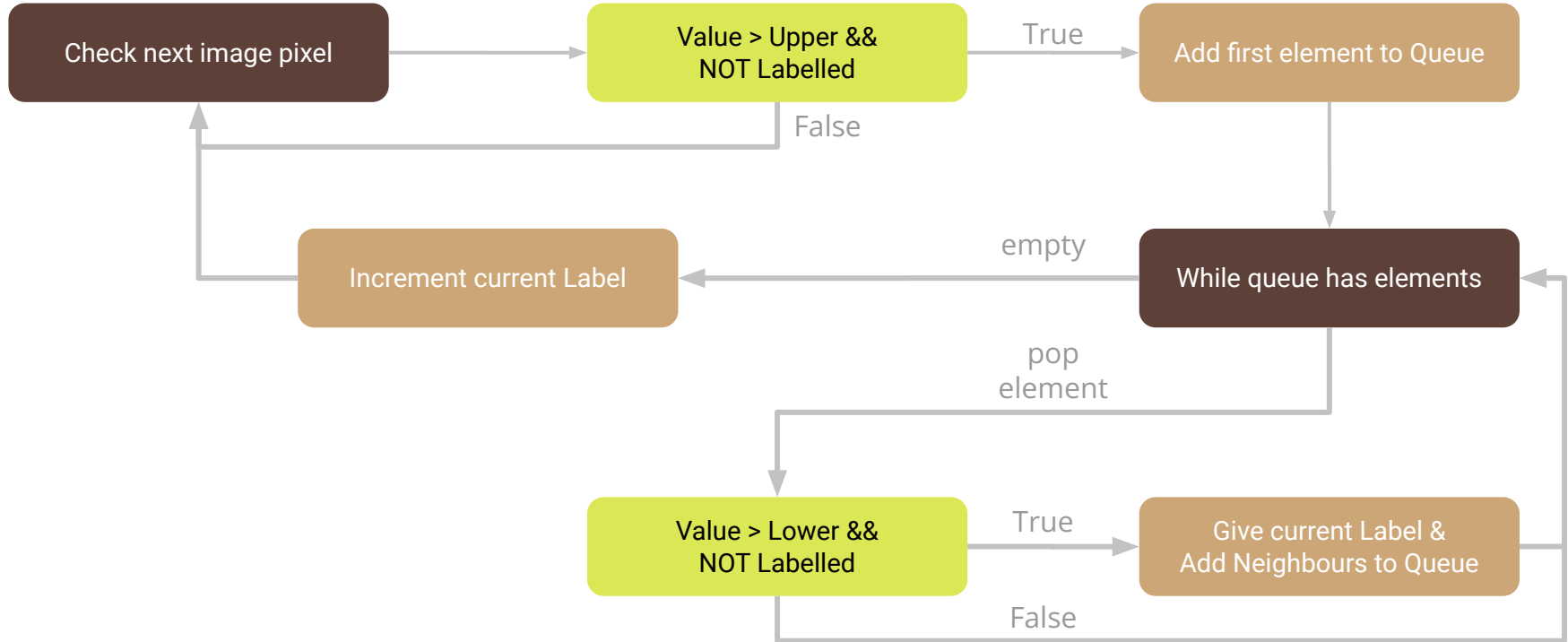


# 8-Way Connectivity



# One Component at a Time Algorithm

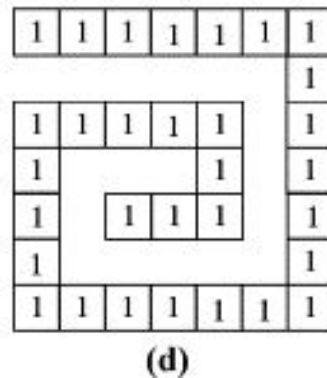
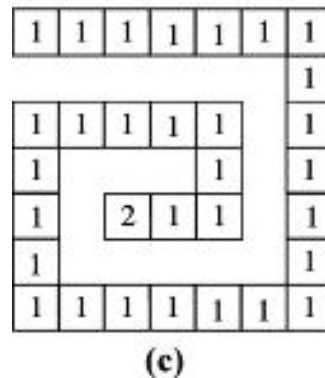
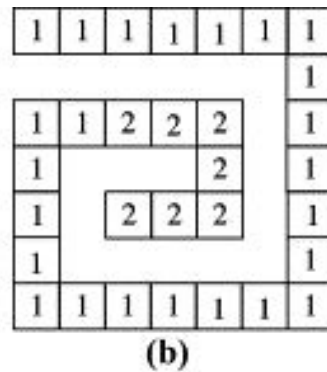
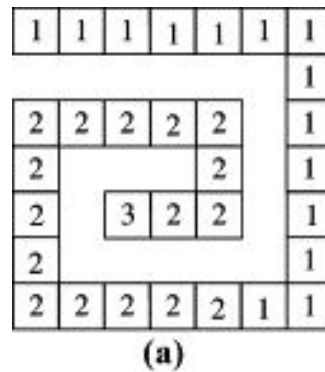
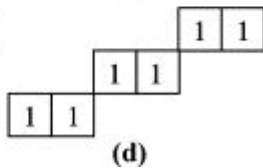
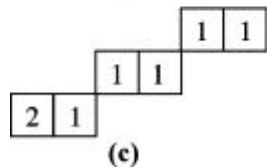
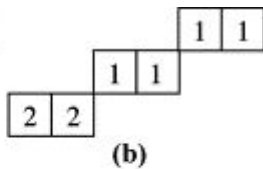
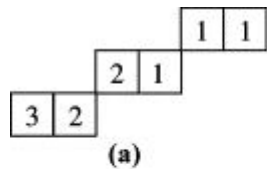
# One Component At a Time



# One Component at a Time

0	0	0	0	1
0	1	0	0	0
1	0	0	1	1
0	0	0	1	0
0	0	0	1	0

## Other variations



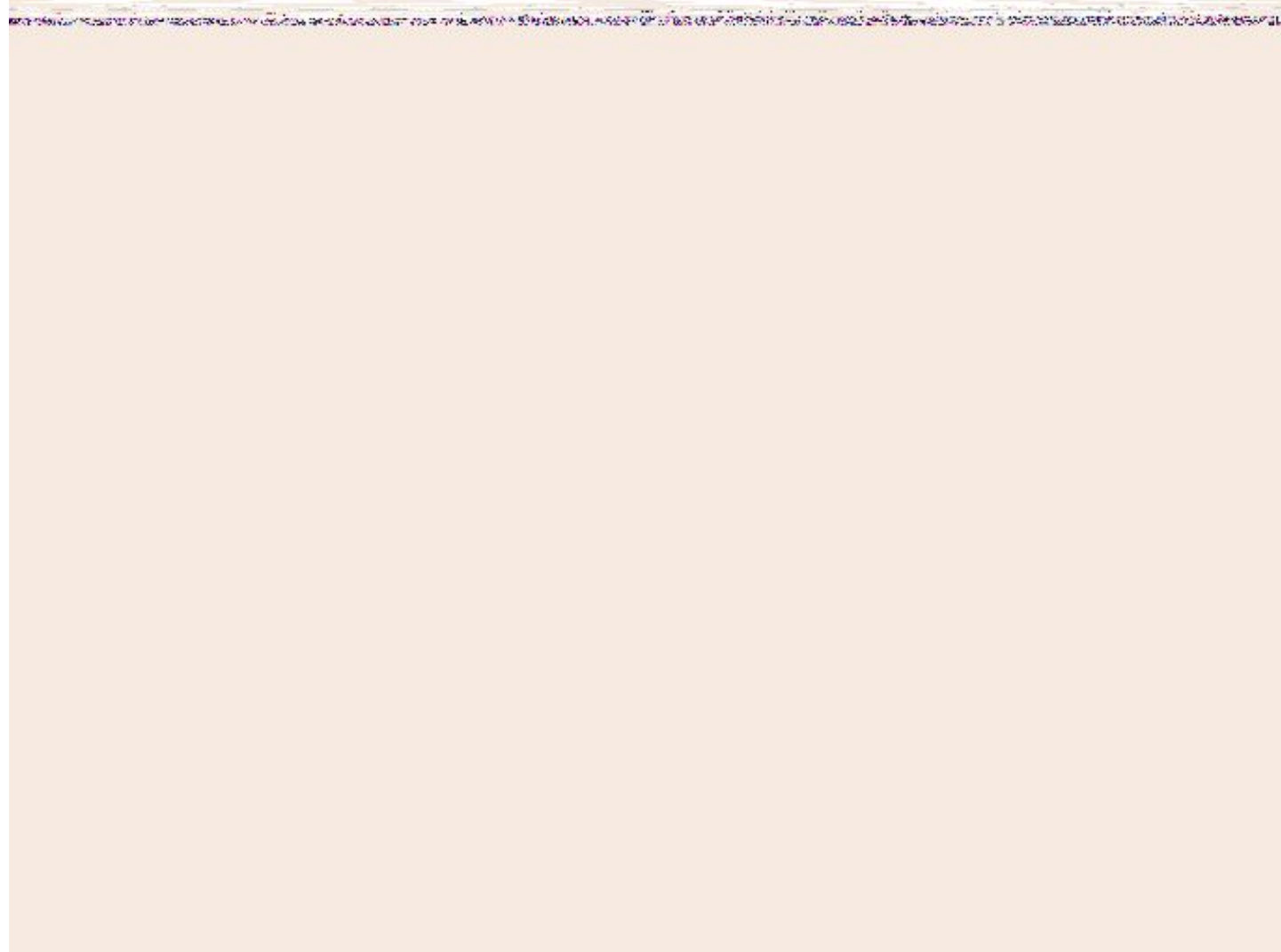
# Two Pass Algorithm

# First Pass

For each non-zero pixel, check its neighbours:

<b>No non-zero neighbours</b>		Give a new Label as it is a new component
<b>One non-zero neighbour</b>		Give the same label
<b>Many non-zero neighbours</b>	<b>Same labels</b>	Give the same label
	<b>Different Labels</b>	Set current pixel to the lowest label Add a note to the equivalences table





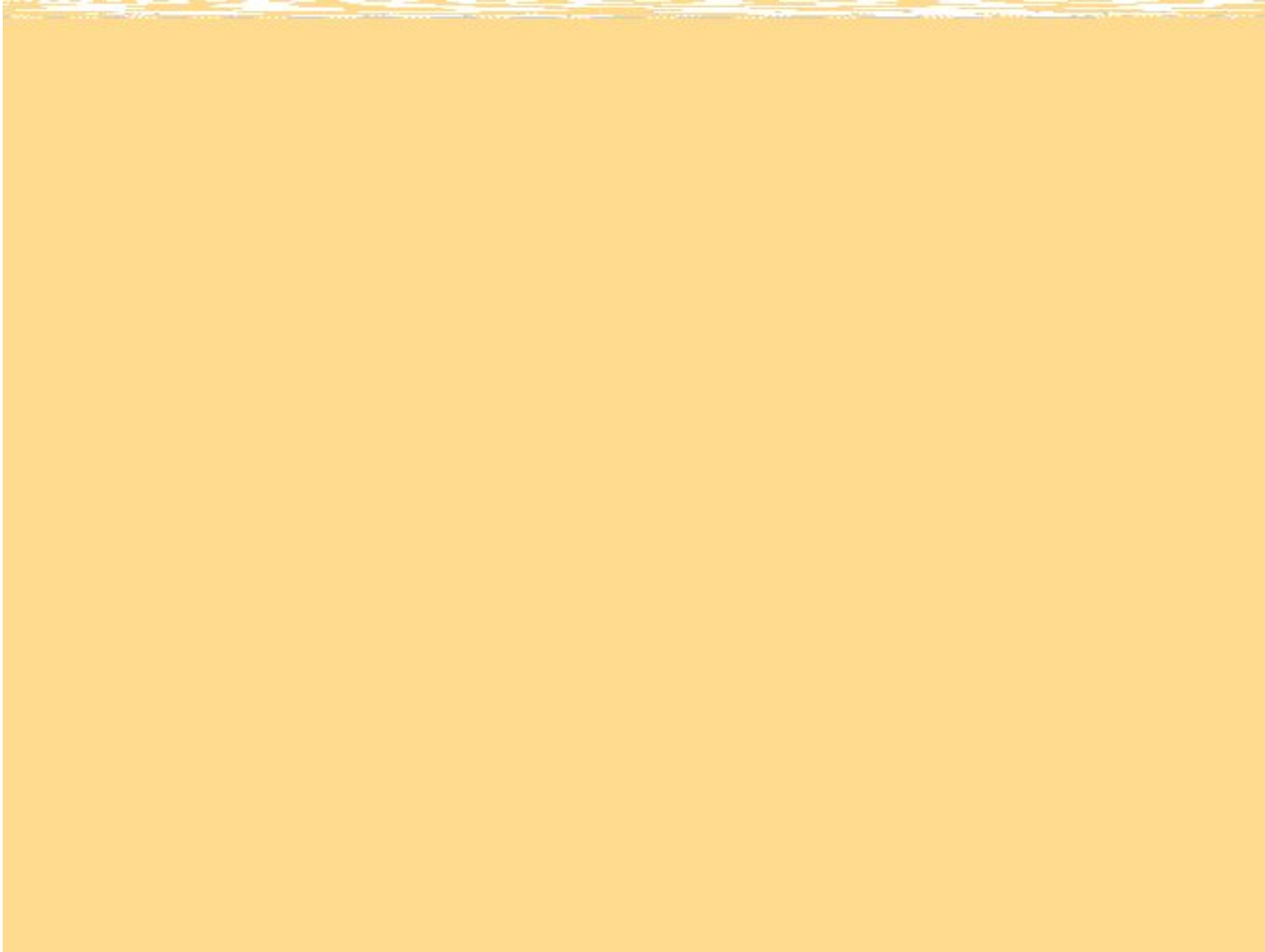
# Second Pass

For each pixel with a label, check the equivalences table and update its value

## Problem

**How to implement this equivalences table?**

- Hashmap
- Disjoint Sets



# Applying to the Edge Detection Problem

A) Thresholded image is not binary

**Naive:** Assume weak and strong pixels as non-zero

**Problem:** Loses weak vs strong information

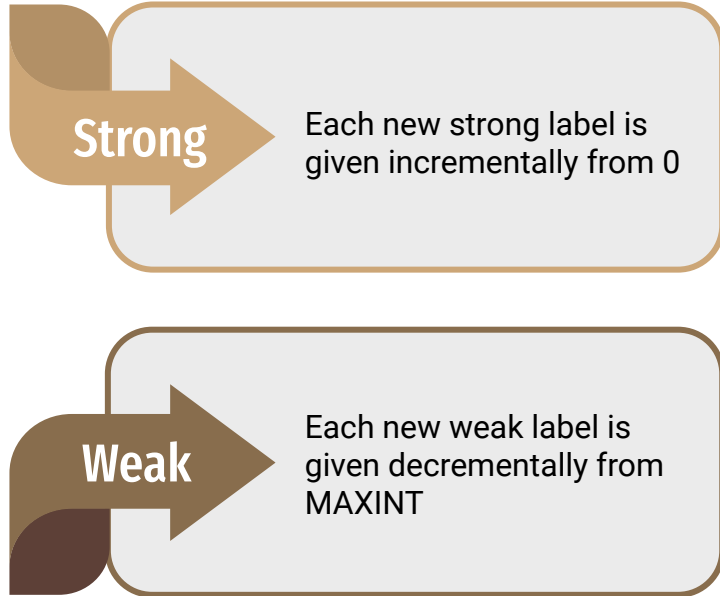
B) Keep only pixels which belong to a strong edge

**Naive:** Third pass to mark if each label is a strong edge

**Problem:** Adding another pass hinders performance

# Proposed Solution: Weak & Strong Labels

Change the label numbering according to the pixel type in the source image



**Weak Label:** Label is higher than the lowest assigned weak label

**Strong Label:** Label is lower than the highest assigned strong label

Updated rules of the first pass:

- When processing a strong pixel, instead of simply copying the label from its neighbours, if the label to be assigned belongs to a weak label, create a new strong label instead and add a new entry to the equivalences table

# 3D Mapping

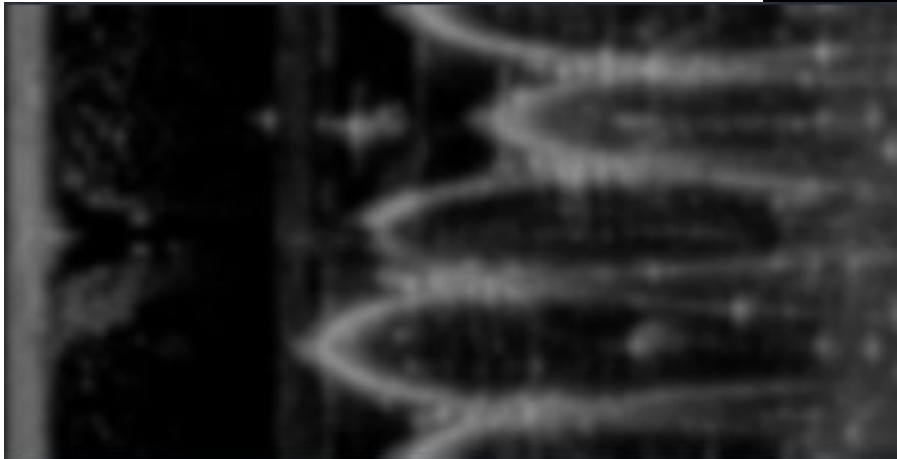
# Range to First Feature

(adapted)

- I. Ignore first few measurements
- ~~II. Apply an intensity threshold~~
- III. Apply Canny's Edge Detector
- IV. Select first non-zero value
- V. Find local Maxima in original

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# Results







Questions?



# References

[0] - João Pedro Bastos Fula - Underwater mapping using a SONAR

[https://en.wikipedia.org/wiki/Canny edge detector](https://en.wikipedia.org/wiki/Canny_edge_detector)

[https://en.wikipedia.org/wiki/Connected-component labeling](https://en.wikipedia.org/wiki/Connected-component_labeling)

[https://en.wikipedia.org/wiki/Deriche edge detector](https://en.wikipedia.org/wiki/Deriche_edge_detector)

[https://moodle.up.pt/pluginfile.php/183979/mod\\_resource/content/13/VCOM\\_04%20-%20Edge%20and%20Line%20Detection.pdf](https://moodle.up.pt/pluginfile.php/183979/mod_resource/content/13/VCOM_04%20-%20Edge%20and%20Line%20Detection.pdf)

<https://towardsdatascience.com/implementing-a-connected-component-labeling-algorithm-from-scratch-94e1636554f>

<http://kiwi.bridgeport.edu/cepeg585/CannyEdgeDetector.pdf>

<https://www.sciencedirect.com/science/article/abs/pii/S1077314202000309?via%3Dihub>

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